COSC 341 Theory of Computing Lecture 18 Propositional logic and satisfiability

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Lecture slides (mostly) by Michael Albert *Keywords*: Propositional logic, satisfiability

Aim: to find our first NP-complete problem

- Finding our first NP-complete problem is daunting as we must be able to reduce *all* problems in NP to it.
- But once we find our first such problem *prob* then any problem we can reduce *prob* to is also NP-complete.
- ► Gradually we can build up a catalogue of NP-complete problems.
- Spoiler: that first NP-complete problem is going to be satisfiability in propositional logic.

Overview of propositional logic (1)

- ▶ VAR is a finite (but large) set of Boolean variable symbols.
- \blacktriangleright \land and \lor are binary operators and \neg is a unary operator.
- ▶ We define a context-free grammar of logical *formulas* over the variables:
 - $E \to V$
 - $E \to E \wedge E$
 - $E \to E \vee E$
 - $E \to \neg E$
 - $V \rightarrow v \;\; (\text{for any} \; v \in \text{VAR})$
- What does $a \wedge b \lor c$ mean?
- Oops! It is ambiguous. We need to add brackets as non-terminals in our grammar:
 - $E \to (V)$

. . .

 $E \to (E) \land (E)$

Overview of propositional logic (2)

• A truth assignment is a map $T : VAR \rightarrow \{t, f\}$.

Given a truth assignment, we define the evaluation of a formula recursively:

 $\begin{aligned} \operatorname{ev}(v,T) &= T(v) \\ \operatorname{ev}(\neg(E),T) &= \operatorname{!ev}(E,T) \text{ where } \operatorname{!t} = \operatorname{f} \text{ and } \operatorname{!f} = \operatorname{t} \\ \operatorname{ev}((E_1) \wedge (E_2),T) &= \begin{cases} \operatorname{t} & \text{if both } \operatorname{ev}(E_1,T) \text{ and } \operatorname{ev}(E_2,T) \text{ are } \operatorname{t} \\ \operatorname{f} & \text{otherwise} \end{cases} \\ \operatorname{ev}((E_1) \vee (E_2),T) &= \begin{cases} \operatorname{f} & \text{if both } \operatorname{ev}(E_1,T) \text{ and } \operatorname{ev}(E_2,T) \text{ are } \operatorname{f} \\ \operatorname{t} & \text{otherwise} \end{cases} \end{aligned}$

► Two formulas are *logically equivalent* if they have the same evaluation for *every* truth assignment, e.g. a ∧ (b ∨ c) and (a ∧ b) ∨ (a ∧ c).

Conjunctive normal form and satisfiability

- A <u>literal</u> (in logic) is a atomic formula or its negation. In propositional logic, the atoms are variables, so the literals are variables and negated variables.
- A <u>clause</u> is a disjunction of literals, e.g.:

 $x_0 \lor x_2 \lor \neg x_6$

A formula in conjunctive normal form (CNF) is a conjunction of clauses, e.g.:

 $(x_0 \lor x_2 \lor \neg x_6) \land (x_1 \lor \neg x_3 \lor x_4 \lor \neg x_6) \land (x_1 \lor x_3 \lor x_5) \land (x_0 \lor \neg x_7)$

- A formula is <u>satisfiable</u> if it evaluates to t for *some* truth assignment T. T is called a <u>satisfying assignment</u>.
- ► The satisfiability problem (slide 8) can take a formula in CNF because:

Theorem

Every formula in propositional logic is logically equivalent to one in CNF.

Proof.

- Construct the truth table for E
- For each row with value f create a clause denying that truth assignment.
- Take the conjunction of those clauses.

Proof illustration

Consider $(q \land \neg p \land \neg r) \lor (p \land r)$.

p	q	r	$(q \land \neg p \land \neg r) \lor (p \land r)$	Created clause
f	f	f	f	$p \lor q \lor r$
f	f	t	f	$p \lor q \lor \neg r$
f	t	f	t	
f	t	t	f	$p \vee \neg q \vee \neg r$
t	f	f	f	$\neg p \lor q \lor r$
t	f	t	t	
t	t	f	f	$\neg p \vee \neg q \vee r$
t	t	t	t	

 $\mathsf{CNF:} \ (p \lor q \lor r) \land (p \lor q \lor \neg r) \land (p \lor \neg q \lor \neg r) \land (\neg p \lor q \lor r) \land (\neg p \lor \neg q \lor r)$

From https://math.stackexchange.com/questions/3549712/how-to-compute-cnf-from-truth-table

(Potential) exponential blow-up

Consider the formula $(x_1 \wedge y_1) \vee (x_2 \wedge y_2) \vee \ldots \vee (x_n \wedge y_n)$.

- This has size linear in n (ignoring the $\log n$ factor for storing variable names).
- Converting this to CNF produces a formula with 2^n clauses:

$$(x_1 \lor x_2 \lor \ldots \lor x_n)$$

$$\land (y_1 \lor x_2 \lor \ldots \lor x_n)$$

$$\land (x_1 \lor y_2 \lor \ldots \lor x_n)$$

$$\land (y_1 \lor y_2 \lor \ldots \lor x_n)$$

$$\ldots$$

$$\land (y_1 \lor y_2 \lor \ldots \lor y_n)$$

- Each clause contains either x_i or y_i (or its negation) for each *i*.
- Can we do better? By counting the number of falsifying assignments needed we can show that there must be at least (1.5)ⁿ clauses.

Satisfiability

Satisfiability or SAT

Instance: A formula in CNF over a set of variables *V*. *Problem*: Does the formula have a satisfying assignment?

- ► The space requirement to describe an instance of SAT containing k clauses over n variables is O(kn log n) (see Notes 16).
- The non-deterministic "guess and check" procedure to verify an instance requires time at most quadratic in the size of the input (see Notes 16).
- So SAT is in NP.

What makes formulas in CNF easy or hard to satisfy?

- A clause with k variables has 2^k truth assignment over those variables.
- How many of these truth assignments make the clause false?
- Only 1. Clauses are hard to falsify. Each disjunct must be false.
- Adding more variables to a clause makes it even harder to falsify / easier to satisfy.
- Adding more clauses to a formula in CNF adds more falsifying truth assignments.
- The greater the number of clauses in a formula in CNF and the shorter those clauses, the more likely the formula is to be unsatisfiable.

SAT is NP-complete (trailer)

How can we possibly reduce *every* NP problem to SAT?

Let a TM, M solve some problem in NP (in polynomial time). As it runs we could imagine taking a snapshot of its configuration at each time step.

- Describing such a snapshot using Boolean variables is pretty straightforward.
- But:
 - How do we ensure the snapshots change correctly from one time step to the next?
 - How do we limit the size given that the TM has an infinite tape?
 - How do we create a formula that is satisfied exactly when M accepts its input?